Heuristic Search: BestFS and A*
No BS Career Success Talk
Date:  Mon., Sept 23
Time:  5:30 pm
Location:  DMP 110

Ericsson Info Session
Date:  Tues., Sept 24
Time:  11:30 am – 1:30 pm
Location:  Kaiser 2020

CS Community Hackathon Info Session
Date:  Tues., Sept 24
Time:  6 pm
Location:  DMP 110

TELUS Open House
Date:  Fri., Sept 27
Time:  12:30 – 3 pm
Location:  3777 Kingsway

IBM Info Session
Date:  Mon., Sept 30
Time:  5:45 – 7:30 pm
Location:  DMP 110

Gameloft Tech Talk
Date:  Tues., Oct 1
Time:  5:30 – 6:30 pm
Location:  DMP 110
Course Announcements

Marks for Assignment0: posted on Connect

Assignment1: posted

If you are confused on basic search algorithm, different search strategies..... Check learning goals at the end of lectures. Work on the Practice Exercises and Please come to office hours

Giuseppe: Fri 2-3, my office CICSR 105
Kamyar Ardekani Mon 2-3, X150 (Learning Center)
Tatsuro Oya Thur 11-12, X150 (Learning Center)
Xin Ru (Nancy) Wang Tue 2-3, X150 (Learning Center)
Course Announcements

Inked Slides

- At the end of each lecture I revise/clean-up the slides. Adding comments, improving writing… make sure you check them out
Lecture Overview

• Recap / Finish Heuristic Function
• Best First Search
• A*
How to Combine Heuristics

If \( h_1(n) \) is admissible and \( h_2(n) \) is also admissible then

A. \( \min( h_1(n), h_2(n) ) \) is also admissible and dominates its components

B. \( \max( h_1(n), h_2(n) ) \) is also admissible and dominates its components

C. \( \text{avg}( h_1(n), h_2(n) ) \) is also admissible and dominates its components

D. \textbf{None} of the above
Example Heuristic Functions

- Another one we can use the number of moves between each tile's current position and its position in the solution.

\[
\begin{array}{ccc}
7 & 2 & 4 \\
5 & 6 & 1 \\
8 & 3 & 1 \\
\end{array}
\quad \begin{array}{ccc}
1 & 2 & 2 \\
3 & 4 & 5 \\
6 & 7 & 8 \\
\end{array}
\quad \begin{array}{ccc}
5 & 4 \\
6 & 1 & 8 \\
7 & 3 & 2 \\
\end{array}
\quad \begin{array}{ccc}
1 & 2 & 3 \\
8 & 4 \\
7 & 6 & 5 \\
\end{array}
\]

Start State \quad \text{Goal State} \quad \text{Start State} \quad \text{Goal State}

Tiles:

\[
\begin{array}{cccccccc}
1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 \\
\end{array}
\]

\[
\begin{array}{cccccccc}
1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 \\
\end{array}
\]

\[
\begin{array}{cccccccc}
3 & 2 & 2 & 2 & 2 & 3 & 3 & 2 \\
\end{array}
\]

\[
\begin{array}{cccccccc}
3 & 2 & 2 & 2 & 2 & 3 & 3 & 2 \\
\end{array}
\]

\[
\begin{array}{cccccccc}
3 & 2 & 2 & 2 & 2 & 3 & 3 & 2 \\
\end{array}
\]

\[
\begin{array}{cccccccc}
= 18 \\
\end{array}
\]
Another approach to construct heuristics

Solution cost for a subproblem

Original Problem

<table>
<thead>
<tr>
<th></th>
<th>1</th>
<th>3</th>
</tr>
</thead>
<tbody>
<tr>
<td>8</td>
<td>2</td>
<td>5</td>
</tr>
<tr>
<td>7</td>
<td>6</td>
<td>4</td>
</tr>
</tbody>
</table>

SubProblem

<table>
<thead>
<tr>
<th></th>
<th>1</th>
<th>3</th>
</tr>
</thead>
<tbody>
<tr>
<td>@</td>
<td>2</td>
<td>@</td>
</tr>
<tr>
<td>@</td>
<td>@</td>
<td>4</td>
</tr>
</tbody>
</table>

Current node

Goal node

CPSC 322, Lecture 3
Combining Heuristics: Example

In 8-puzzle, solution cost for the 1,2,3,4 subproblem is substantially more accurate than sum of Manhattan distance of each tile from its goal position in some cases.

So.....
Admissible heuristic for Vacuum world?

**States?** Where it is dirty and robot location

**Actions?** *Left*, *Right*, *Suck*

**Possible goal test?** no dirt at all locations
Admissible heuristic for Vacuum world?

states? Where it is dirty and robot location
actions? Left, Right, Suck
Possible goal test? no dirt at all locations
Lecture Overview

• Recap Heuristic Function
• Best First Search
• A*
Best-First Search

- **Idea:** select the path whose end is closest to a goal according to the heuristic function.

- **Best-First search** selects a path on the frontier with minimal \( h \)-value (for the end node).

- It treats the frontier as a priority queue ordered by \( h \).
  (similar to CFS by cost)

- This is a greedy approach: it always takes the path which appears locally best
Analysis of Best-First Search

- **Not Complete**: a low heuristic value can mean that a cycle gets followed forever.

- **Optimal**: no (why not?)
- **Time complexity** is $O(b^m)$
- **Space complexity** is $O(b^m)$
Lecture Overview

- Recap Heuristic Function
- Best First Search
- A* Search Strategy
How can we effectively use $h(n)$

Maybe we should combine it with the cost. How?
Shall we select from the frontier the path $\rho$ with:

A. Lowest $\text{cost}(\rho) - h(\rho)$
B. Highest $\text{cost}(\rho) - h(\rho)$
C. Highest $\text{cost}(\rho) + h(\rho)$
D. Lowest $\text{cost}(\rho) + h(\rho)$
\[ A^* \text{ Search Algorithm} \]

- \( A^* \) is a mix of:
  - lowest-cost-first and
  - best-first search

- \( A^* \) treats the frontier as a priority queue ordered by \( f(p) = \text{cost}(p) + h(p) \), is an estimate.

- It always selects the node on the frontier with the lowest estimated total distance.
Computing f-values

f-value of UBC → KD → JB? 6 9 10 11
Analysis of A*

If the heuristic is completely uninformative and the edge costs are all the same, A* is equivalent to….

A. BFS
B. LCFS
C. DFS
D. None of the Above
Analysis of $A^*$

Let's assume that arc costs are strictly positive.

- **Time complexity** is $O(b^m)$ as $h(s) = 0$, for all states.
  - the heuristic could be completely uninformative and the edge costs could all be the same, meaning that $A^*$ does the same thing as DFS, BFS, LCFS.

- **Space complexity** is $O(b^m)$ like BFS, $A^*$ maintains a frontier which grows with the size of the tree.

- **Completeness**: yes.

- **Optimality**: ??
Optimality of $A^*$

If $A^*$ returns a solution, that solution is guaranteed to be optimal, as long as

When

- the branching factor is finite
- arc costs are strictly positive
- $h(n)$ is an underestimate of the length of the shortest path from $n$ to a goal node, and is non-negative

**Theorem**

If $A^*$ selects a path $p$ as the solution, $p$ is the shortest (i.e., lowest-cost) path.
Why is $A^*$ optimal?

- $A^*$ returns $p$
- Assume for contradiction that some other path $p'$ is actually the shortest path to a goal
- Consider the moment when $p$ is chosen from the frontier. Some part of path $p'$ will also be on the frontier; let's call this partial path $p''$.

Think: for any path from start to any state, there is always a subpath of that path on the frontier.
Why is $A^*$ optimal? (cont’)

- Because $p$ was expanded before $p''$,
- Because $p$ is a goal, $h(p) = 0$ Thus
- Because $h$ is admissible, $\text{cost}(p'') + h(p'') \leq \text{cost}(p)$ for any path $p'$ to a goal that extends $p''$
  
  Thus $\text{cost}(p) \leq \text{cost}(p')$ for any other path $p'$ to a goal.

This contradicts our assumption that $p'$ is the shortest path.
In fact, we can prove something even stronger about $A^*$: in a sense (given the particular heuristic that is available) no search algorithm could do better!

Optimal Efficiency: Among all optimal algorithms that start from the same start node and use the same heuristic $h$, $A^*$ expands the minimal number of paths.
Sample A* applications

• An Efficient A* Search Algorithm For Statistical Machine Translation. 2001


  • Machine Vision … Here we consider a new compositional model for finding salient curves.

• Factored A* search for models over sequences and trees International Conference on AI. 2003….
  It starts saying… The primary challenge when using A* search is to find heuristic functions that simultaneously are admissible, close to actual completion costs, and efficient to calculate… applied to NLP and Bioinformatics
Sample A* applications (cont’)

DFS, BFS, A* Animation Example

• The AI-Search animation system

• To examine Search strategies when they are applied to the 8puzzle

• Compare only DFS, BFS and A* (with only the two heuristics we saw in class)

  • With default start state and goal
  • DFS will find Solution at depth 32
  • BFS will find Optimal solution depth 6
  • A* will also find opt. sol. expanding much less nodes
nPuzzles are not always solvable

Half of the starting positions for the \( n \)-puzzle are impossible to resolve (for more info on 8puzzle) http://www.isle.org/~sbay/ics171/project/unsolvable

- So experiment with the AI-Search animation system with the default configurations.
- If you want to try new ones keep in mind that you may pick unsolvable problems
Learning Goals for today’s class

• Define/read/write/trace/debug & Compare different search algorithms
  • With / Without cost
  • Informed / Uninformed

• Formally prove A* optimality.
Next class

Finish Search  (finish Chpt 3)
- Branch-and-Bound
- A* enhancements
- Non-heuristic Pruning
- Dynamic Programming